# An Overview of the DDMA team

Kevin R. Vixie

The Data Driven Modeling and Analysis (DDMA) Team &
Mathematical Modeling and Analysis (T-7)
Los Alamos National Laboratory

September 28, 2006





#### The team started in 2002

- Essential Component Analysis Proposal: eventually 2 of 3 proposals were funded.
- 2002 Image Analysis Workshop at LANL: high visibility, good attendance
- Initial members: Vixie (lead), Asaki, Wohlberg

#### Since that time we have grown

- 25 members: roughly half at LANL, half at universities
- Many parallel projects, huge range of expertise
- 2.5M\$/year
- Innovative team structure and activities





# DDMA – Data-Driven Modeling and Analysis

Data analysis requires a synthesis of expertise from many fields.

#### Lead Team

Tom Asaki (CCS-2) asaki@lanl.gov

Katharine Chartrand (HPC-4) kncx@lanl.gov

Rick Chartrand (T-7) rickc@lanl.gov

Matt Sottile (CCS-1) matt@lanl.gov

Kevin Vixie (T-7) vixie@lanl.gov

#### Team Members

Bill Allard (Duke)

Erik Bollt (Clarkson) Patrick Campbell (T-7)

David Caraballo (Georgetown)

John Dennis, Jr. (Rice)

David Dreisigmever (HPC-4)

Selim Esedoglu (Michigan)

Gilad Lerman (Minnesota)

Fred Park (Michigan)

Collin Powell (CCS-2)

Brvan Rasmussen (HPC-4)

Paul Rodriguez (T-7)

Pete Schultz (Clarkson)

Valentina Staneva (T-7)

Curt Vogel (Montana State)

Brendt Wohlberg (T-7)

Wotao Yin (Rice)

Mark Abramson (AFIT)

David Arathorn (General Intelligence Corp.)

Chris Orum (D-1)

Robert Sarracino (Scientific Modeling Group, Los Alamos)

#### Expertise

Algorithmics

Computational Science

Analysis

Geometric Measure Theory

Partial Differential Equations

Variational Analysis Harmonic Analysis

Differential Geometry

**Dynamical Systems** 

Dimension Reduction

Inverse Problems Numerical Analysis

Optimization

Signal Processing

Statistics

Tomography

#### **Current Applications**

Algorithm and Prototype Software Development

Comparison Metrics Dimension Reduction

Extrapolation and Inpainting

Feature Measures

Mixed-Variable Optimization Object Recognition

Special X-ray Tomography Warping Transformations

Geometric Image Processing





## Geometric Analysis

- Image Warping
- Applied Differential Geometry
- Radiographic Inversions
- Object Detection/recognition/segmentation
- Metrics and Regularizations
- Compressed Sensing
- Optimization
- Other areas





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- Numerous Special Sessions: SIAM Imaging, SIAM Dynamical Systems, AMS meetings and AAAS Annual Meeting
- ▶ 2002 LANL Image Analysis Workshop
- ▶ 2005 Graduate Summer School at IPAM-UCLA
- RIPS Summer program at IPAM: last 3 years
- Numerous Short Courses and Invited Talks
- ▶ DDMA Speaker Series: Over 40 speakers in the first 1.5 years.
- ▶ ( ... and half the team is at universities!)





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- Current Students: Campbell, Staneva, Powell
- ► So far, 6 postdocs and 14 students





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- External members are key team members, not simply collaborators on papers. Students are often key contributers, projects are not simply invented and tailored to fit them.
- We convert all classified problems to unclassified analogs and move the problems to any appropriate member of the team, irrespective of clearance level.
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- Sparse X-ray Inversions: Fast and Cheap (and in control)
- Non-convex Regularizations: More from Less
- Using Geometric Measure Theory: Properties of Exact Minimizers
- Metrics that ignore the unimportant: Null-Riemannian Geometry
- Image Warping: Monge-Kantorovich and others
- Random Shapes and Images: important for many reasons
- Object detection and recognition: not just faces





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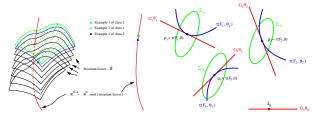




## Classification Mod Invariance: Face Results

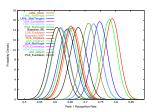
DDMA -

## Early Project: Ignoring differences that don't matter



Factor out the unimportant

Tangent Approximations to Orbits





Results: CSU database (2003)

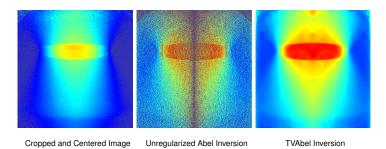


## TVAbel: Inversions assuming symmetry

Another of the earliest DDMA projects:

$$\min_{u} F(u) = \int |\nabla u| dx + \lambda \int |Pu - d|^2 dx \tag{1}$$

where the P is the measurement operator (Abel projection).





# IDA: The DDMA Image and Data Analyzer

IDA provides a single framework for executing complex analysis pipelines. It has advantages over plain Matlab:

- Consistent interface for all components, regardless of language.
- Automatic data management via an embedded SQL database.
- Basic visualization capabilities.
- A simple data flow language is provided.
- Built in "parameter study" capability.
- Abstraction of iteration allows for transparent parallel/distributed execution, data-reuse.

IDA makes analysis easier, and allows contributors to rapidly make their components available by conforming to our component standard.





## Member Research: Examples

DDMA

In the next two slides I show a very short snapshot from a team member's research.

- Wotao Yin was selected because he responded quickly to a request for a synopsis.
- As mentioned above, this is a small slice of the full breadth and depth of the team.

You will hear four of us talk throughout the day about different aspects of our research.





## DDMA Member Research: Wotao Yin (Rice)

- ► Convex optimization theory and applications, especially the applications of second-order cone programming and semi-definite programming in image processing, statistical learning, and computer vision.
- Extensions of some existing optimization algorithms to sovling problems constrained on smooth manifolds.
- Very recent work includes a fast algorithm for minimizing certain convex functions using the parametric maximum flow algorithm.





# DDMA Member Research: Wotao Yin (Rice)

#### Using L1TV to remove lighting effects:



Faces under lighting.jpg

